

whale

May 19, 2019

```
In [8]: import numpy as np
import pandas as pd
import os
import matplotlib.pyplot as plt
import matplotlib.image as mpling
from matplotlib.pyplot import imshow
from keras.utils import plot_model
from PIL import Image as pil_image

from sklearn.preprocessing import LabelEncoder
from sklearn.preprocessing import OneHotEncoder

from keras import layers
from keras.preprocessing import image
from keras.applications.imagenet_utils import preprocess_input
from keras.layers import Input, Dense, Activation, BatchNormalization, Flatten, Conv2D
from keras.layers import AveragePooling2D, MaxPooling2D, Dropout
from keras.models import Model

import keras.backend as K
from keras.models import Sequential

import warnings
warnings.simplefilter("ignore", category=DeprecationWarning)

In [2]: train_df = pd.read_csv("~/Downloads/humpback-whale-identification/train.csv")
train_df.head()

Out[2]:
```

	Image	Id
0	0000e88ab.jpg	w_f48451c
1	0001f9222.jpg	w_c3d896a
2	00029d126.jpg	w_20df2c5
3	00050a15a.jpg	new_whale
4	0005c1ef8.jpg	new_whale

```
In [6]: print("Preparing images")
X_train = np.zeros((train_df.shape[0], 128, 128, 3))
count = 0
```

```

for fig in train_df['Image']:
    #load images into images of size 128x128x3
    img = image.load_img("train/"+fig, target_size=(128, 128, 3))
    x = image.img_to_array(img)
    x = preprocess_input(x)

    X_train[count] = x
    if (count%500 == 0):
        print("Processing image: ", count+1, ", ", fig)
    count += 1
X_train = X_train / 255.0
print(X_train.shape)

```

Preparing images

```

Processing image: 1 , 0000e88ab.jpg
Processing image: 501 , 04c72257b.jpg
Processing image: 1001 , 09cacb84d.jpg
Processing image: 1501 , 0ef961892.jpg
Processing image: 2001 , 141b56a1a.jpg
Processing image: 2501 , 199a417aa.jpg
Processing image: 3001 , 1ec170983.jpg
Processing image: 3501 , 23f084b93.jpg
Processing image: 4001 , 29163ad0b.jpg
Processing image: 4501 , 2e0fab120.jpg
Processing image: 5001 , 3347515d9.jpg
Processing image: 5501 , 3842d71dc.jpg
Processing image: 6001 , 3d7f4c7d5.jpg
Processing image: 6501 , 425f763ca.jpg
Processing image: 7001 , 4714400cd.jpg
Processing image: 7501 , 4c082fbdf.jpg
Processing image: 8001 , 50c683e23.jpg
Processing image: 8501 , 560d986ad.jpg
Processing image: 9001 , 5b68c83ed.jpg
Processing image: 9501 , 60410f111.jpg
Processing image: 10001 , 654951f81.jpg
Processing image: 10501 , 6a572256c.jpg
Processing image: 11001 , 6f96f55b6.jpg
Processing image: 11501 , 74da2b511.jpg
Processing image: 12001 , 7989d9a27.jpg
Processing image: 12501 , 7e5aa2d8a.jpg
Processing image: 13001 , 832382cfb.jpg
Processing image: 13501 , 87f6c0a15.jpg
Processing image: 14001 , 8cfc22e5d.jpg
Processing image: 14501 , 91dcfedcd.jpg
Processing image: 15001 , 97079398e.jpg
Processing image: 15501 , 9c2ad64a9.jpg
Processing image: 16001 , a11956dff.jpg

```

```
Processing image: 16501 , a5f9ffe86.jpg
Processing image: 17001 , aaf1a967b.jpg
Processing image: 17501 , af9a1ffc6.jpg
Processing image: 18001 , b4e02531d.jpg
Processing image: 18501 , ba2355ca6.jpg
Processing image: 19001 , bf60e7fed.jpg
Processing image: 19501 , c49f39ce3.jpg
Processing image: 20001 , c960111d0.jpg
Processing image: 20501 , ce7984d8a.jpg
Processing image: 21001 , d38efaec9.jpg
Processing image: 21501 , d831d28ee.jpg
Processing image: 22001 , dd3ca2387.jpg
Processing image: 22501 , e288d66cf.jpg
Processing image: 23001 , e7cc793db.jpg
Processing image: 23501 , ec8c7229d.jpg
Processing image: 24001 , f1b850552.jpg
Processing image: 24501 , f6af8a4b8.jpg
Processing image: 25001 , fc09f2302.jpg
(25361, 128, 128, 3)
```

```
In [7]: for i in range(10):
        plt.imshow(X_train[i][:,:,0], cmap="gray")
        plt.title(plt.title(train_df.iloc[i,0]))
        plt.axis("off")
        plt.show()
```

Text(0.5, 1.0, '0000e88ab.jpg')



Text(0.5, 1.0, '0001f9222.jpg')



Text(0.5, 1.0, '00029d126.jpg')



Text(0.5, 1.0, '00050a15a.jpg')



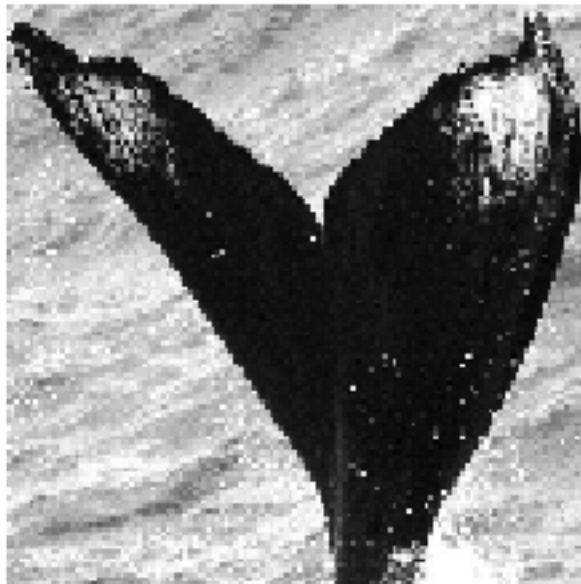
Text(0.5, 1.0, '0005c1ef8.jpg')



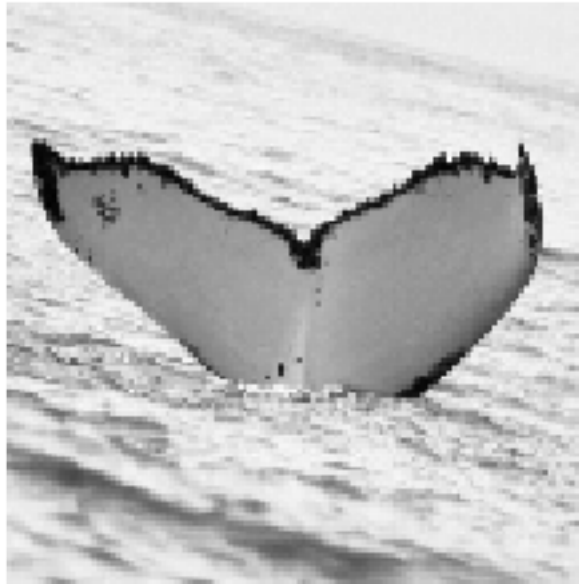
Text(0.5, 1.0, '0006e997e.jpg')



Text(0.5, 1.0, '000a6daec.jpg')



Text(0.5, 1.0, '000f0f2bf.jpg')



Text(0.5, 1.0, '0016b897a.jpg')



Text(0.5, 1.0, '001c1ac5f.jpg')



```
In [8]: values = np.array(train_df['Id'])
label_encoder = LabelEncoder()
integer_encoded = label_encoder.fit_transform(values)
# print(integer_encoded)
onehot_encoder = OneHotEncoder(sparse=False)
integer_encoded = integer_encoded.reshape(len(integer_encoded), 1)
onehot_encoded = onehot_encoder.fit_transform(integer_encoded)
# print(onehot_encoded)
y = onehot_encoded
print(y.shape)
```

(25361, 5005)

0.1 Distribution of images per whale is highly skewed.

2000+ whales have just one image

Single whale with most images have 73 of them

Images dsitribution:

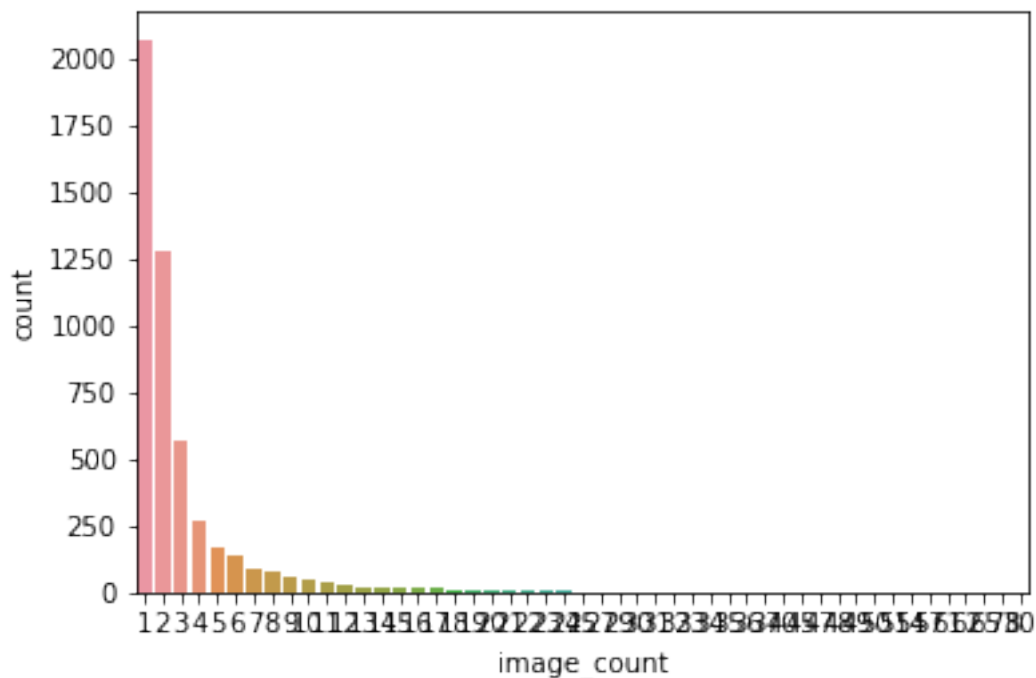
- almost 30% comes from whales with 4 or less images
- almost 40% comes from 'new_whale' group
- the rest 30% comes from whales with 5-73 images

```
In [21]: train_df['Id'].value_counts()[:4]
```



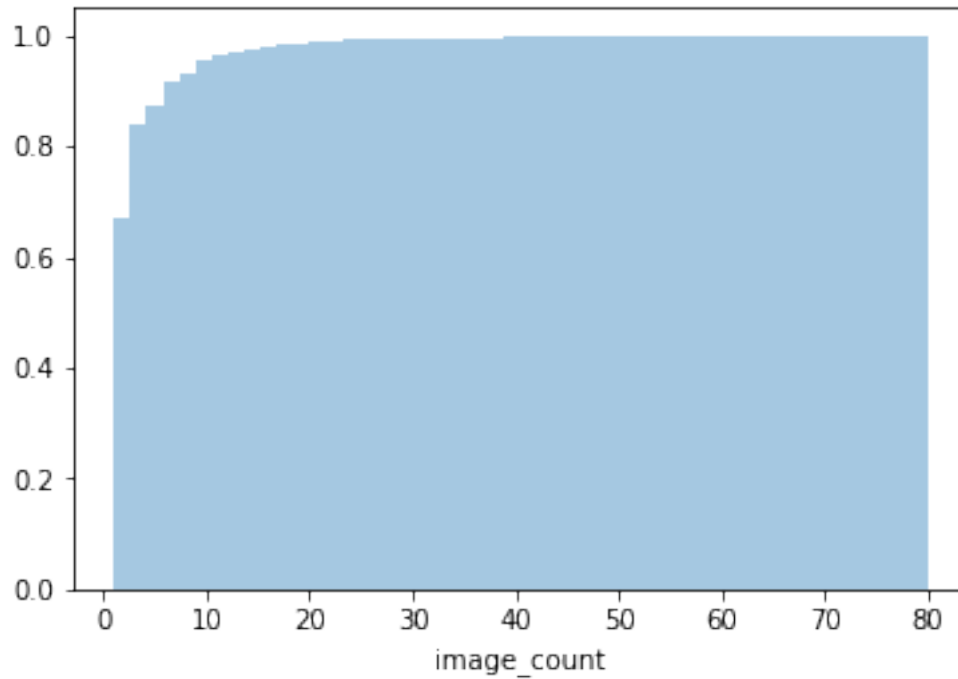
```
Out[21]: new_whale      9664
         w_23a388d       73
         w_9b5109b       65
         w_9c506f6       62
         Name: Id, dtype: int64
```

```
In [23]: import seaborn as sns
counted = train_df.groupby("Id").count().rename(columns={"Image": "image_count"})
counted.loc[counted["image_count"] > 80, 'image_count'] = 80
plt.figure()
sns.countplot(data=counted, x="image_count")
plt.show()
sns.distplot(counted["image_count"], norm_hist=True, kde=False, hist_kws={'cumulative
```



```
/anaconda3/lib/python3.6/site-packages/matplotlib/axes/_axes.py:6521: MatplotlibDeprecationWarning:
The 'normed' kwarg was deprecated in Matplotlib 2.1 and will be removed in 3.1. Use 'density' :
alternative="density", removal="3.1")
```

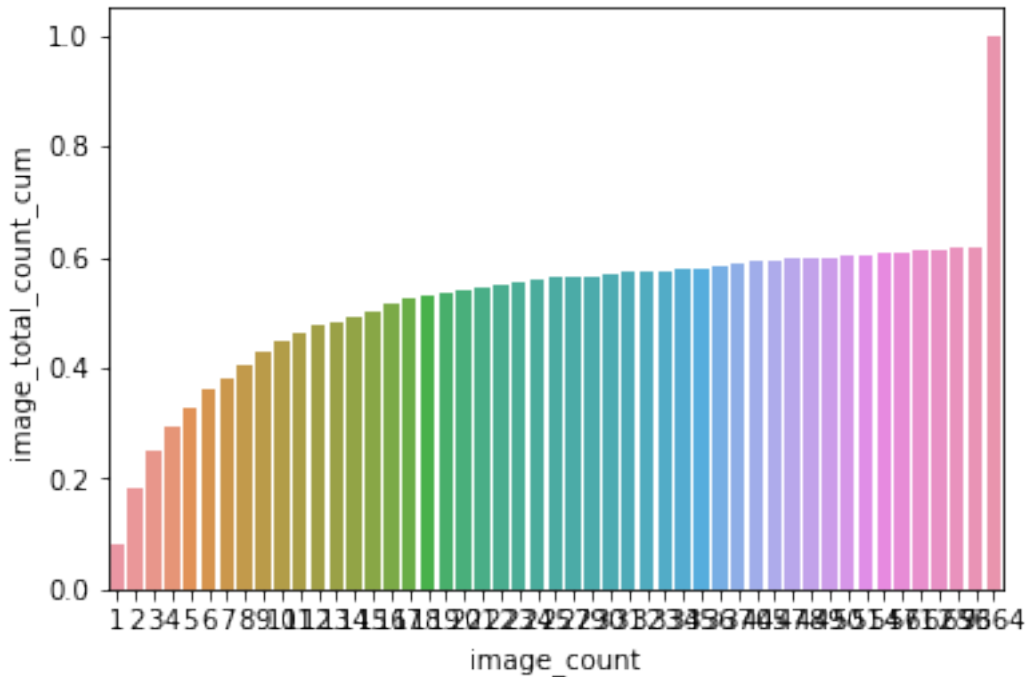
```
Out[23]: <matplotlib.axes._subplots.AxesSubplot at 0x1a1c118c18>
```



This shows that 2000 of the 5K classes have only one image. About 1250 classes have 2 images, and so on.

```
In [26]: image_count_for_whale = train_df.groupby("Id", as_index=False).count().rename(columns:
whale_count_for_image_count = image_count_for_whale.groupby("image_count", as_index=False)
whale_count_for_image_count['image_total_count'] = whale_count_for_image_count['image
whale_count_for_image_count['image_total_count_cum'] = whale_count_for_image_count["i
sns.barplot(x='image_count',y='image_total_count_cum',data=whale_count_for_image_count
```

```
Out[26]: <matplotlib.axes._subplots.AxesSubplot at 0x1de607a0f0>
```



```
In [27]: whale_count_for_image_count[:5]
```

```
Out[27]:
```

	image_count	whale_count	image_total_count	image_total_count_cum
0	1	2073	2073	0.081740
1	2	1285	2570	0.183076
2	3	568	1704	0.250266
3	4	273	1092	0.293324
4	5	172	860	0.327235

```
In [28]: whale_count_for_image_count[-3:]
```

```
Out[28]:
```

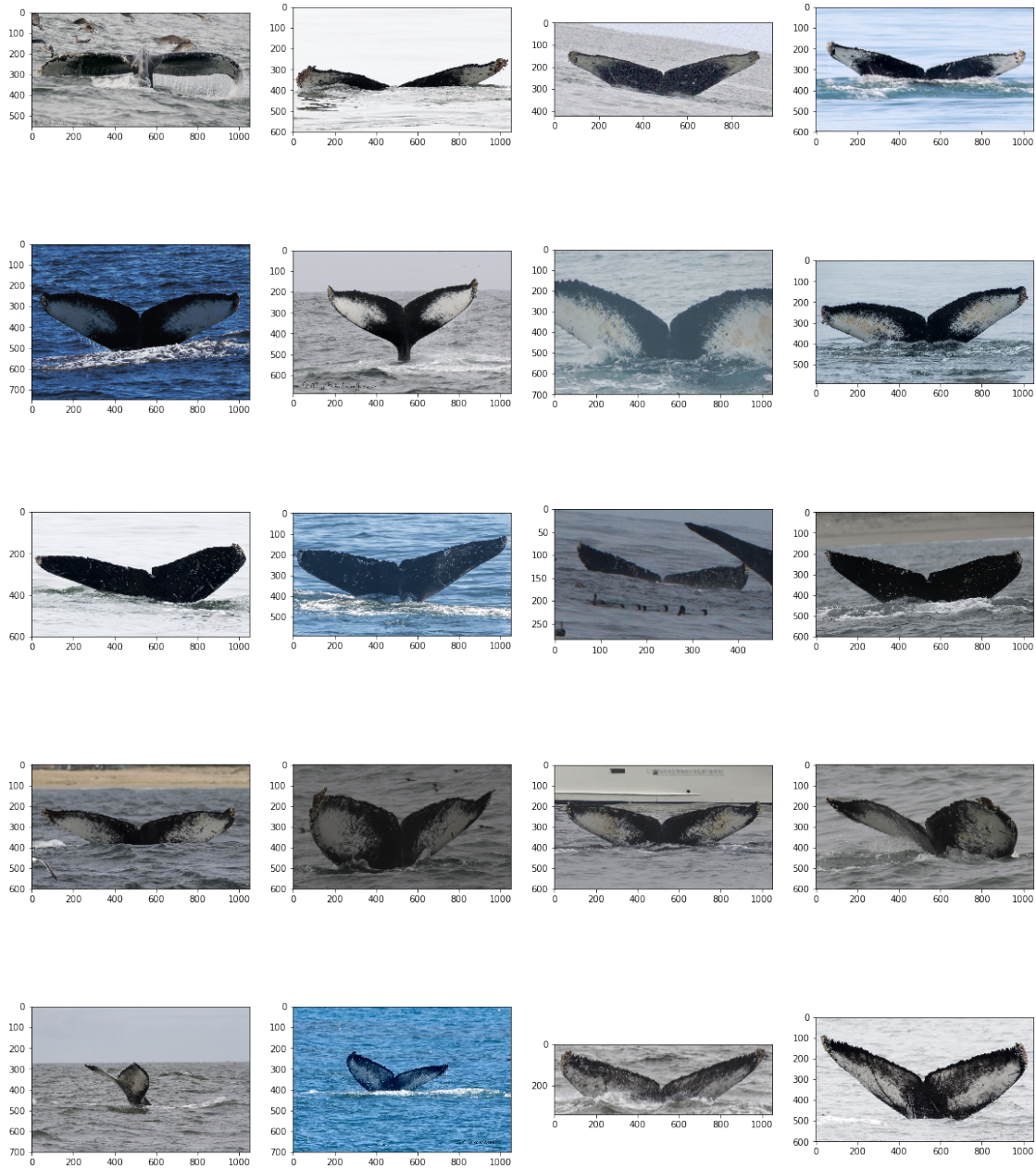
	image_count	whale_count	image_total_count	image_total_count_cum
46	65	1	65	0.616064
47	73	1	73	0.618942
48	9664	1	9664	1.000000

```
In [33]: import cv2
topN=5
top_whales = train_df['Id'].value_counts().index[1:1+topN]
fig = plt.figure(figsize = (20, 5*topN))

for widx, whale in enumerate(top_whales):
    for idx, img_name in enumerate(train_df[train_df['Id'] == whale]['Image'][:4]):
        axes = widx*4 + idx+1
        y = fig.add_subplot(topN, 4, axes)
```

```
img = cv2.imread(os.path.join("", "train", img_name))
img = cv2.cvtColor(img, cv2.COLOR_BGR2RGB)
y.imshow(img)
```

```
plt.show()
```



over 7000 unique resolutions but 39 most popular cover ~45% images (both in train and in test)

```
In [5]: import collections
import os
```

```

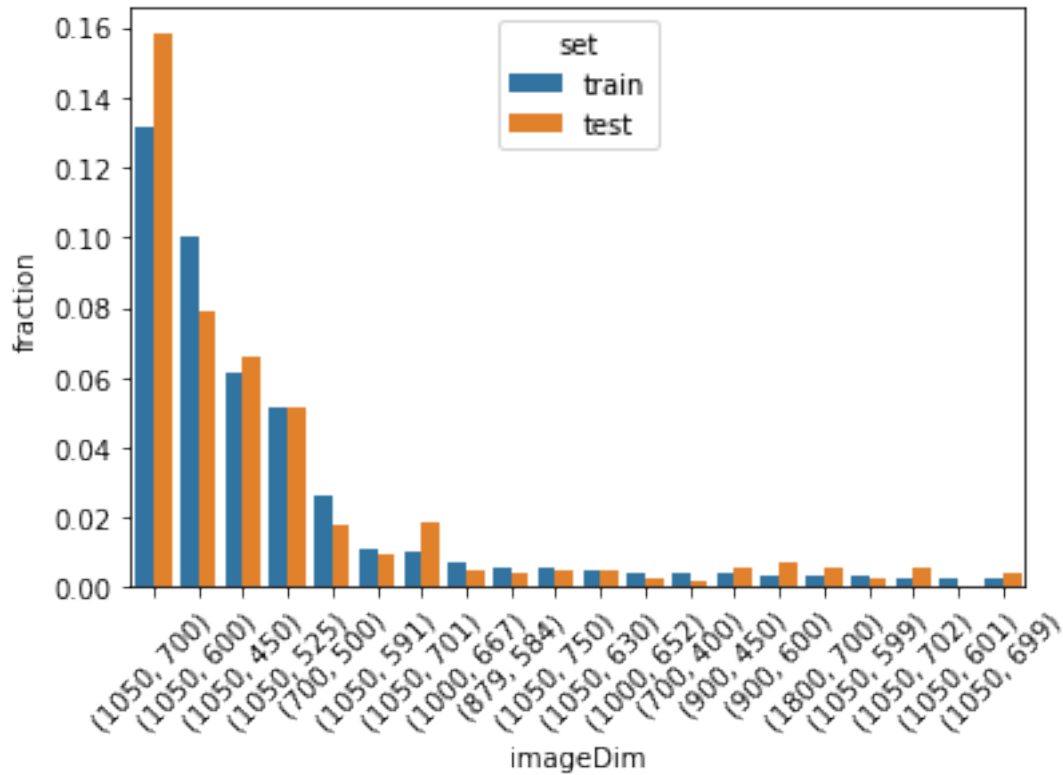
from PIL import Image
imageSizes_train = collections.Counter([Image.open(f'train/{filename}').size
                                         for filename in os.listdir(f"train")])
imageSizes_test = collections.Counter([Image.open(f'test/{filename}').size
                                       for filename in os.listdir(f"test")])

In [11]: import seaborn as sns
def isdf(imageSizes):
    imageSizeFrame = pd.DataFrame(list(imageSizes.most_common()), columns = ["imageDim", "count"])
    imageSizeFrame['fraction'] = imageSizeFrame['count'] / sum(imageSizes.values())
    imageSizeFrame['count_cum'] = imageSizeFrame['count'].cumsum()
    imageSizeFrame['count_cum_fraction'] = imageSizeFrame['count_cum'] / sum(imageSizes.values())
    return imageSizeFrame

train_isdf = isdf(imageSizes_train)
train_isdf['set'] = 'train'
test_isdf = isdf(imageSizes_test)
test_isdf['set'] = 'test'

isizes = train_isdf.merge(test_isdf, how="outer", on="imageDim")
isizes['total_count'] = isizes['count_x'] + isizes['count_y']
dims_order = isizes.sort_values('total_count', ascending=False)[['imageDim']]
len(dims_order)
isizes = pd.concat([train_isdf, test_isdf])
popularSizes = isizes[isizes['fraction'] > 0.002]
popularSizes.groupby('set').max()['count_cum_fraction']
sns.barplot(x='imageDim', y='fraction', data = popularSizes, hue="set")
_ = plt.xticks(rotation=45)

```



1 Siamese Neural Network

```
In [26]: from keras import regularizers
from keras.optimizers import Adam
from keras.engine.topology import Input
from keras.layers import Activation, Add, BatchNormalization, Concatenate, Conv2D, Dense
from keras.models import Model

img_shape = (384,384,1) # The image shape used by the model

def subblock(x, filter, **kwargs):
    x = BatchNormalization()(x)
    y = x
    y = Conv2D(filter, (1, 1), activation='relu', **kwargs)(y) # Reduce the number of
    y = BatchNormalization()(y)
    y = Conv2D(filter, (3, 3), activation='relu', **kwargs)(y) # Extend the feature f
    y = BatchNormalization()(y)
    y = Conv2D(K.int_shape(x)[-1], (1, 1), **kwargs)(y) # no activation # Restore the
    y = Add()([x,y]) # Add the bypass connection
    y = Activation('relu')(y)
    return y
```

```

def build_model(lr, l2, activation='sigmoid'):

    #####
    # BRANCH MODEL
    #####
    regul = regularizers.l2(l2)
    optim = Adam(lr=lr)
    kwargs = {'padding':'same', 'kernel_regularizer':regul}

    inp = Input(shape=img_shape) # 384x384x1
    x = Conv2D(64, (9,9), strides=2, activation='relu', **kwargs)(inp)

    x = MaxPooling2D((2, 2), strides=(2, 2))(x) # 96x96x64
    for _ in range(2):
        x = BatchNormalization()(x)
        x = Conv2D(64, (3,3), activation='relu', **kwargs)(x)

    x = MaxPooling2D((2, 2), strides=(2, 2))(x) # 48x48x64
    x = BatchNormalization()(x)
    x = Conv2D(128, (1,1), activation='relu', **kwargs)(x) # 48x48x128
    for _ in range(4): x = subblock(x, 64, **kwargs)

    x = MaxPooling2D((2, 2), strides=(2, 2))(x) # 24x24x128
    x = BatchNormalization()(x)
    x = Conv2D(256, (1,1), activation='relu', **kwargs)(x) # 24x24x256
    for _ in range(4): x = subblock(x, 64, **kwargs)

    x = MaxPooling2D((2, 2), strides=(2, 2))(x) # 12x12x256
    x = BatchNormalization()(x)
    x = Conv2D(384, (1,1), activation='relu', **kwargs)(x) # 12x12x384
    for _ in range(4): x = subblock(x, 96, **kwargs)

    x = MaxPooling2D((2, 2), strides=(2, 2))(x) # 6x6x384
    x = BatchNormalization()(x)
    x = Conv2D(512, (1,1), activation='relu', **kwargs)(x) # 6x6x512
    for _ in range(4): x = subblock(x, 128, **kwargs)

    x = GlobalMaxPooling2D()(x) # 512
    branch_model = Model(inp, x)

    #####
    # HEAD MODEL
    #####
    mid = 32
    xa_inp = Input(shape=branch_model.output_shape[1:])
    xb_inp = Input(shape=branch_model.output_shape[1:])
    x1 = Lambda(lambda x : x[0]*x[1])([xa_inp, xb_inp])

```

```

x2          = Lambda(lambda x : x[0] + x[1])([xa_inp, xb_inp])
x3          = Lambda(lambda x : K.abs(x[0] - x[1]))([xa_inp, xb_inp])
x4          = Lambda(lambda x : K.square(x))(x3)
x           = Concatenate()(x1, x2, x3, x4)
x           = Reshape((4, branch_model.output_shape[1], 1), name='reshape1')(x)

# Per feature NN with shared weight is implemented using CONV2D with appropriate
x           = Conv2D(mid, (4, 1), activation='relu', padding='valid')(x)
x           = Reshape((branch_model.output_shape[1], mid, 1))(x)
x           = Conv2D(1, (1, mid), activation='linear', padding='valid')(x)
x           = Flatten(name='flatten')(x)

# Weighted sum implemented as a Dense layer.
x           = Dense(1, use_bias=True, activation=activation, name='weighted-average')
head_model = Model([xa_inp, xb_inp], x, name='head')

#####
# SIAMESE NEURAL NETWORK
#####
# Complete model is constructed by calling the branch model on each input image,
# and then the head model on the resulting 512-vectors.
img_a       = Input(shape=img_shape)
img_b       = Input(shape=img_shape)
xa          = branch_model(img_a)
xb          = branch_model(img_b)
x           = head_model([xa, xb])
model       = Model([img_a, img_b], x)
model.compile(optimizer, loss='binary_crossentropy', metrics=['binary_crossentropy'],
return model, branch_model, head_model

model, branch_model, head_model = build_model(64e-5,0)
head_model.summary()

```

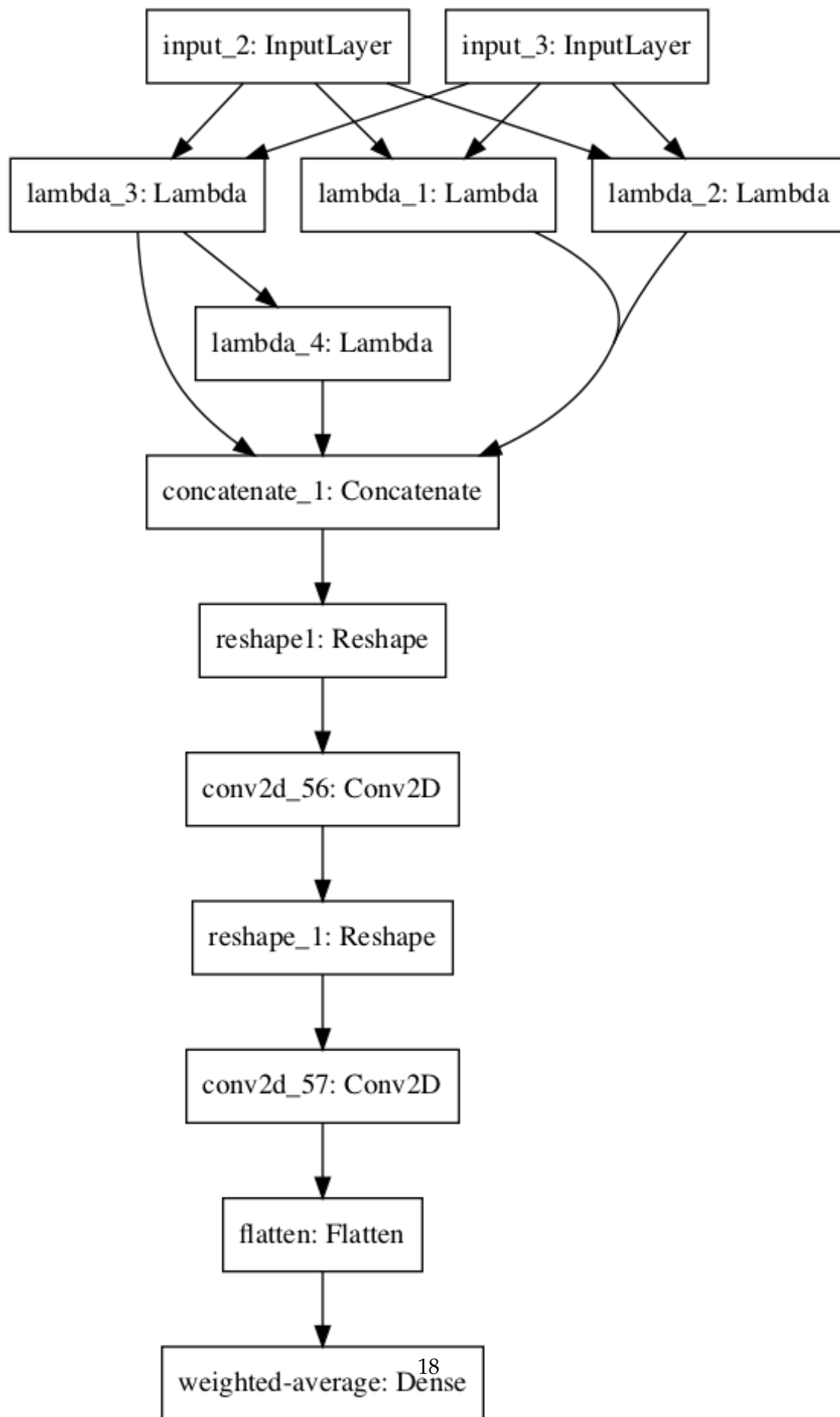
Layer (type)	Output Shape	Param #	Connected to
input_2 (InputLayer)	(None, 512)	0	
input_3 (InputLayer)	(None, 512)	0	
lambda_3 (Lambda)	(None, 512)	0	input_2[0][0] input_3[0][0]
lambda_1 (Lambda)	(None, 512)	0	input_2[0][0] input_3[0][0]
lambda_2 (Lambda)	(None, 512)	0	input_2[0][0] input_3[0][0]

lambda_4 (Lambda)	(None, 512)	0	lambda_3[0][0]
concatenate_1 (Concatenate)	(None, 2048)	0	lambda_1[0][0] lambda_2[0][0] lambda_3[0][0] lambda_4[0][0]
reshape1 (Reshape)	(None, 4, 512, 1)	0	concatenate_1[0][0]
conv2d_56 (Conv2D)	(None, 1, 512, 32)	160	reshape1[0][0]
reshape_1 (Reshape)	(None, 512, 32, 1)	0	conv2d_56[0][0]
conv2d_57 (Conv2D)	(None, 512, 1, 1)	33	reshape_1[0][0]
flatten (Flatten)	(None, 512)	0	conv2d_57[0][0]
weighted-average (Dense)	(None, 1)	513	flatten[0][0]

=====
 Total params: 706
 Trainable params: 706
 Non-trainable params: 0
 =====

```
In [27]: plot_model(head_model, to_file='head-model.png')
         pil_image.open('head-model.png')
```

```
Out[27]:
```



```
In [28]: branch_model.summary()
```

Layer (type)	Output Shape	Param #	Connected to
input_1 (InputLayer)	(None, 384, 384, 1)	0	
conv2d_1 (Conv2D)	(None, 192, 192, 64)	5248	input_1[0][0]
max_pooling2d_1 (MaxPooling2D)	(None, 96, 96, 64)	0	conv2d_1[0][0]
batch_normalization_1 (BatchNor	(None, 96, 96, 64)	256	max_pooling2d_1[0][0]
conv2d_2 (Conv2D)	(None, 96, 96, 64)	36928	batch_normalization_1[0][0]
batch_normalization_2 (BatchNor	(None, 96, 96, 64)	256	conv2d_2[0][0]
conv2d_3 (Conv2D)	(None, 96, 96, 64)	36928	batch_normalization_2[0][0]
max_pooling2d_2 (MaxPooling2D)	(None, 48, 48, 64)	0	conv2d_3[0][0]
batch_normalization_3 (BatchNor	(None, 48, 48, 64)	256	max_pooling2d_2[0][0]
conv2d_4 (Conv2D)	(None, 48, 48, 128)	8320	batch_normalization_3[0][0]
batch_normalization_4 (BatchNor	(None, 48, 48, 128)	512	conv2d_4[0][0]
conv2d_5 (Conv2D)	(None, 48, 48, 64)	8256	batch_normalization_4[0][0]
batch_normalization_5 (BatchNor	(None, 48, 48, 64)	256	conv2d_5[0][0]
conv2d_6 (Conv2D)	(None, 48, 48, 64)	36928	batch_normalization_5[0][0]
batch_normalization_6 (BatchNor	(None, 48, 48, 64)	256	conv2d_6[0][0]
conv2d_7 (Conv2D)	(None, 48, 48, 128)	8320	batch_normalization_6[0][0]
add_1 (Add)	(None, 48, 48, 128)	0	batch_normalization_4[0][0] conv2d_7[0][0]
activation_1 (Activation)	(None, 48, 48, 128)	0	add_1[0][0]
batch_normalization_7 (BatchNor	(None, 48, 48, 128)	512	activation_1[0][0]
conv2d_8 (Conv2D)	(None, 48, 48, 64)	8256	batch_normalization_7[0][0]

batch_normalization_8 (BatchNor	(None, 48, 48, 64)	256	conv2d_8[0][0]
conv2d_9 (Conv2D)	(None, 48, 48, 64)	36928	batch_normalization_8[0][0]
batch_normalization_9 (BatchNor	(None, 48, 48, 64)	256	conv2d_9[0][0]
conv2d_10 (Conv2D)	(None, 48, 48, 128)	8320	batch_normalization_9[0][0]
add_2 (Add)	(None, 48, 48, 128)	0	batch_normalization_7[0][0] conv2d_10[0][0]
activation_2 (Activation)	(None, 48, 48, 128)	0	add_2[0][0]
batch_normalization_10 (BatchNo	(None, 48, 48, 128)	512	activation_2[0][0]
conv2d_11 (Conv2D)	(None, 48, 48, 64)	8256	batch_normalization_10[0][0]
batch_normalization_11 (BatchNo	(None, 48, 48, 64)	256	conv2d_11[0][0]
conv2d_12 (Conv2D)	(None, 48, 48, 64)	36928	batch_normalization_11[0][0]
batch_normalization_12 (BatchNo	(None, 48, 48, 64)	256	conv2d_12[0][0]
conv2d_13 (Conv2D)	(None, 48, 48, 128)	8320	batch_normalization_12[0][0]
add_3 (Add)	(None, 48, 48, 128)	0	batch_normalization_10[0][0] conv2d_13[0][0]
activation_3 (Activation)	(None, 48, 48, 128)	0	add_3[0][0]
batch_normalization_13 (BatchNo	(None, 48, 48, 128)	512	activation_3[0][0]
conv2d_14 (Conv2D)	(None, 48, 48, 64)	8256	batch_normalization_13[0][0]
batch_normalization_14 (BatchNo	(None, 48, 48, 64)	256	conv2d_14[0][0]
conv2d_15 (Conv2D)	(None, 48, 48, 64)	36928	batch_normalization_14[0][0]
batch_normalization_15 (BatchNo	(None, 48, 48, 64)	256	conv2d_15[0][0]
conv2d_16 (Conv2D)	(None, 48, 48, 128)	8320	batch_normalization_15[0][0]
add_4 (Add)	(None, 48, 48, 128)	0	batch_normalization_13[0][0] conv2d_16[0][0]
activation_4 (Activation)	(None, 48, 48, 128)	0	add_4[0][0]

max_pooling2d_3 (MaxPooling2D)	(None, 24, 24, 128)	0	activation_4[0][0]
batch_normalization_16 (BatchNo	(None, 24, 24, 128)	512	max_pooling2d_3[0][0]
conv2d_17 (Conv2D)	(None, 24, 24, 256)	33024	batch_normalization_16[0][0]
batch_normalization_17 (BatchNo	(None, 24, 24, 256)	1024	conv2d_17[0][0]
conv2d_18 (Conv2D)	(None, 24, 24, 64)	16448	batch_normalization_17[0][0]
batch_normalization_18 (BatchNo	(None, 24, 24, 64)	256	conv2d_18[0][0]
conv2d_19 (Conv2D)	(None, 24, 24, 64)	36928	batch_normalization_18[0][0]
batch_normalization_19 (BatchNo	(None, 24, 24, 64)	256	conv2d_19[0][0]
conv2d_20 (Conv2D)	(None, 24, 24, 256)	16640	batch_normalization_19[0][0]
add_5 (Add)	(None, 24, 24, 256)	0	batch_normalization_17[0][0] conv2d_20[0][0]
activation_5 (Activation)	(None, 24, 24, 256)	0	add_5[0][0]
batch_normalization_20 (BatchNo	(None, 24, 24, 256)	1024	activation_5[0][0]
conv2d_21 (Conv2D)	(None, 24, 24, 64)	16448	batch_normalization_20[0][0]
batch_normalization_21 (BatchNo	(None, 24, 24, 64)	256	conv2d_21[0][0]
conv2d_22 (Conv2D)	(None, 24, 24, 64)	36928	batch_normalization_21[0][0]
batch_normalization_22 (BatchNo	(None, 24, 24, 64)	256	conv2d_22[0][0]
conv2d_23 (Conv2D)	(None, 24, 24, 256)	16640	batch_normalization_22[0][0]
add_6 (Add)	(None, 24, 24, 256)	0	batch_normalization_20[0][0] conv2d_23[0][0]
activation_6 (Activation)	(None, 24, 24, 256)	0	add_6[0][0]
batch_normalization_23 (BatchNo	(None, 24, 24, 256)	1024	activation_6[0][0]
conv2d_24 (Conv2D)	(None, 24, 24, 64)	16448	batch_normalization_23[0][0]
batch_normalization_24 (BatchNo	(None, 24, 24, 64)	256	conv2d_24[0][0]
conv2d_25 (Conv2D)	(None, 24, 24, 64)	36928	batch_normalization_24[0][0]

batch_normalization_25 (BatchNo	(None, 24, 24, 64)	256	conv2d_25[0][0]
conv2d_26 (Conv2D)	(None, 24, 24, 256)	16640	batch_normalization_25[0][0]
add_7 (Add)	(None, 24, 24, 256)	0	batch_normalization_23[0][0] conv2d_26[0][0]
activation_7 (Activation)	(None, 24, 24, 256)	0	add_7[0][0]
batch_normalization_26 (BatchNo	(None, 24, 24, 256)	1024	activation_7[0][0]
conv2d_27 (Conv2D)	(None, 24, 24, 64)	16448	batch_normalization_26[0][0]
batch_normalization_27 (BatchNo	(None, 24, 24, 64)	256	conv2d_27[0][0]
conv2d_28 (Conv2D)	(None, 24, 24, 64)	36928	batch_normalization_27[0][0]
batch_normalization_28 (BatchNo	(None, 24, 24, 64)	256	conv2d_28[0][0]
conv2d_29 (Conv2D)	(None, 24, 24, 256)	16640	batch_normalization_28[0][0]
add_8 (Add)	(None, 24, 24, 256)	0	batch_normalization_26[0][0] conv2d_29[0][0]
activation_8 (Activation)	(None, 24, 24, 256)	0	add_8[0][0]
max_pooling2d_4 (MaxPooling2D)	(None, 12, 12, 256)	0	activation_8[0][0]
batch_normalization_29 (BatchNo	(None, 12, 12, 256)	1024	max_pooling2d_4[0][0]
conv2d_30 (Conv2D)	(None, 12, 12, 384)	98688	batch_normalization_29[0][0]
batch_normalization_30 (BatchNo	(None, 12, 12, 384)	1536	conv2d_30[0][0]
conv2d_31 (Conv2D)	(None, 12, 12, 96)	36960	batch_normalization_30[0][0]
batch_normalization_31 (BatchNo	(None, 12, 12, 96)	384	conv2d_31[0][0]
conv2d_32 (Conv2D)	(None, 12, 12, 96)	83040	batch_normalization_31[0][0]
batch_normalization_32 (BatchNo	(None, 12, 12, 96)	384	conv2d_32[0][0]
conv2d_33 (Conv2D)	(None, 12, 12, 384)	37248	batch_normalization_32[0][0]
add_9 (Add)	(None, 12, 12, 384)	0	batch_normalization_30[0][0] conv2d_33[0][0]
activation_9 (Activation)	(None, 12, 12, 384)	0	add_9[0][0]

batch_normalization_33 (BatchNo	(None, 12, 12, 384)	1536	activation_9[0][0]
conv2d_34 (Conv2D)	(None, 12, 12, 96)	36960	batch_normalization_33[0][0]
batch_normalization_34 (BatchNo	(None, 12, 12, 96)	384	conv2d_34[0][0]
conv2d_35 (Conv2D)	(None, 12, 12, 96)	83040	batch_normalization_34[0][0]
batch_normalization_35 (BatchNo	(None, 12, 12, 96)	384	conv2d_35[0][0]
conv2d_36 (Conv2D)	(None, 12, 12, 384)	37248	batch_normalization_35[0][0]
add_10 (Add)	(None, 12, 12, 384)	0	batch_normalization_33[0][0] conv2d_36[0][0]
activation_10 (Activation)	(None, 12, 12, 384)	0	add_10[0][0]
batch_normalization_36 (BatchNo	(None, 12, 12, 384)	1536	activation_10[0][0]
conv2d_37 (Conv2D)	(None, 12, 12, 96)	36960	batch_normalization_36[0][0]
batch_normalization_37 (BatchNo	(None, 12, 12, 96)	384	conv2d_37[0][0]
conv2d_38 (Conv2D)	(None, 12, 12, 96)	83040	batch_normalization_37[0][0]
batch_normalization_38 (BatchNo	(None, 12, 12, 96)	384	conv2d_38[0][0]
conv2d_39 (Conv2D)	(None, 12, 12, 384)	37248	batch_normalization_38[0][0]
add_11 (Add)	(None, 12, 12, 384)	0	batch_normalization_36[0][0] conv2d_39[0][0]
activation_11 (Activation)	(None, 12, 12, 384)	0	add_11[0][0]
batch_normalization_39 (BatchNo	(None, 12, 12, 384)	1536	activation_11[0][0]
conv2d_40 (Conv2D)	(None, 12, 12, 96)	36960	batch_normalization_39[0][0]
batch_normalization_40 (BatchNo	(None, 12, 12, 96)	384	conv2d_40[0][0]
conv2d_41 (Conv2D)	(None, 12, 12, 96)	83040	batch_normalization_40[0][0]
batch_normalization_41 (BatchNo	(None, 12, 12, 96)	384	conv2d_41[0][0]
conv2d_42 (Conv2D)	(None, 12, 12, 384)	37248	batch_normalization_41[0][0]
add_12 (Add)	(None, 12, 12, 384)	0	batch_normalization_39[0][0]

			conv2d_42[0][0]
activation_12 (Activation)	(None, 12, 12, 384)	0	add_12[0][0]
max_pooling2d_5 (MaxPooling2D)	(None, 6, 6, 384)	0	activation_12[0][0]
batch_normalization_42 (BatchNo	(None, 6, 6, 384)	1536	max_pooling2d_5[0][0]
conv2d_43 (Conv2D)	(None, 6, 6, 512)	197120	batch_normalization_42[0][0]
batch_normalization_43 (BatchNo	(None, 6, 6, 512)	2048	conv2d_43[0][0]
conv2d_44 (Conv2D)	(None, 6, 6, 128)	65664	batch_normalization_43[0][0]
batch_normalization_44 (BatchNo	(None, 6, 6, 128)	512	conv2d_44[0][0]
conv2d_45 (Conv2D)	(None, 6, 6, 128)	147584	batch_normalization_44[0][0]
batch_normalization_45 (BatchNo	(None, 6, 6, 128)	512	conv2d_45[0][0]
conv2d_46 (Conv2D)	(None, 6, 6, 512)	66048	batch_normalization_45[0][0]
add_13 (Add)	(None, 6, 6, 512)	0	batch_normalization_43[0][0] conv2d_46[0][0]
activation_13 (Activation)	(None, 6, 6, 512)	0	add_13[0][0]
batch_normalization_46 (BatchNo	(None, 6, 6, 512)	2048	activation_13[0][0]
conv2d_47 (Conv2D)	(None, 6, 6, 128)	65664	batch_normalization_46[0][0]
batch_normalization_47 (BatchNo	(None, 6, 6, 128)	512	conv2d_47[0][0]
conv2d_48 (Conv2D)	(None, 6, 6, 128)	147584	batch_normalization_47[0][0]
batch_normalization_48 (BatchNo	(None, 6, 6, 128)	512	conv2d_48[0][0]
conv2d_49 (Conv2D)	(None, 6, 6, 512)	66048	batch_normalization_48[0][0]
add_14 (Add)	(None, 6, 6, 512)	0	batch_normalization_46[0][0] conv2d_49[0][0]
activation_14 (Activation)	(None, 6, 6, 512)	0	add_14[0][0]
batch_normalization_49 (BatchNo	(None, 6, 6, 512)	2048	activation_14[0][0]
conv2d_50 (Conv2D)	(None, 6, 6, 128)	65664	batch_normalization_49[0][0]

batch_normalization_50 (BatchNo	(None, 6, 6, 128)	512	conv2d_50[0][0]
conv2d_51 (Conv2D)	(None, 6, 6, 128)	147584	batch_normalization_50[0][0]
batch_normalization_51 (BatchNo	(None, 6, 6, 128)	512	conv2d_51[0][0]
conv2d_52 (Conv2D)	(None, 6, 6, 512)	66048	batch_normalization_51[0][0]
add_15 (Add)	(None, 6, 6, 512)	0	batch_normalization_49[0][0] conv2d_52[0][0]
activation_15 (Activation)	(None, 6, 6, 512)	0	add_15[0][0]
batch_normalization_52 (BatchNo	(None, 6, 6, 512)	2048	activation_15[0][0]
conv2d_53 (Conv2D)	(None, 6, 6, 128)	65664	batch_normalization_52[0][0]
batch_normalization_53 (BatchNo	(None, 6, 6, 128)	512	conv2d_53[0][0]
conv2d_54 (Conv2D)	(None, 6, 6, 128)	147584	batch_normalization_53[0][0]
batch_normalization_54 (BatchNo	(None, 6, 6, 128)	512	conv2d_54[0][0]
conv2d_55 (Conv2D)	(None, 6, 6, 512)	66048	batch_normalization_54[0][0]
add_16 (Add)	(None, 6, 6, 512)	0	batch_normalization_52[0][0] conv2d_55[0][0]
activation_16 (Activation)	(None, 6, 6, 512)	0	add_16[0][0]
global_max_pooling2d_1 (GlobalM	(None, 512)	0	activation_16[0][0]

=====

Total params: 2,692,096
Trainable params: 2,674,304
Non-trainable params: 17,792

```
In [29]: plot_model(branch_model, to_file='branch-model.png')
         img = pil_image.open('branch-model.png')
         img.resize([x//2 for x in img.size])
```

Out[29]:



